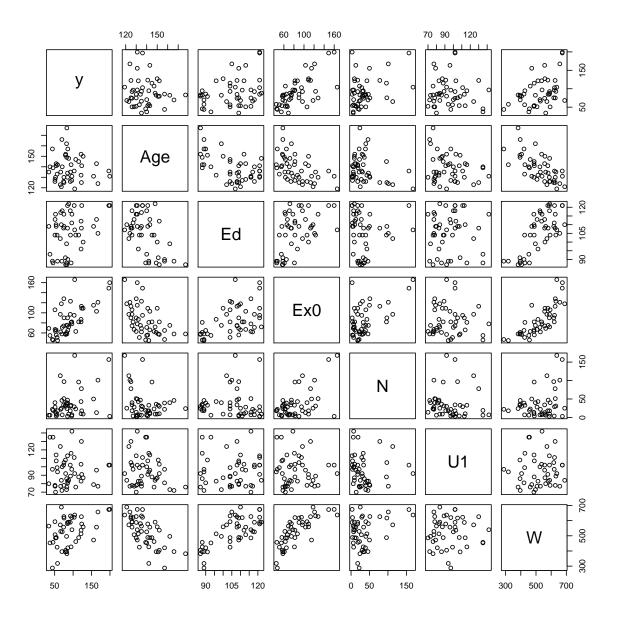
## STAT 401: R example for sum of squares

## Zepu Zhang November 4, 2010

I found a US crime dataset on the internet (likely similar to one of the datasets mentioned in the textbook). Description is attached.

```
> data <- read.table('USCrime.txt', header = TRUE)</pre>
> print(names(data))
 [1] "R"
           "Age" "S"
                       "Ed" "Ex0" "Ex1" "LF" "M" "N" "NW" "U1" "U2"
           "X"
[13] "W"
> # 'R' is the reponse (crime rate)
> # Since we know some predictors are highly correlated,
> # and we don't deal with that problem now,
> # we'll focus on the following predictors:
> # Age: number of male aged 14--24 per 1000 population
> # Ed: mean # of years of schooling
> # Ex0: per capita expenditure on police by government
          state population
> # U1: unemployment rate of urban males
          median family goods
> # W:
>
> y <- data$R
> X <- as.matrix(data[, c('Age', 'Ed', 'Ex0', 'N', 'U1', 'W')])</pre>
> X <- cbind(1, X)
                    # Add the constant predictor.
> colnames(X)[1] <- 'Const'</pre>
> n \leftarrow length(y)
> p <- ncol(X)
> alpha <- .05
> ##### BLOCK 1 #####
> # Take a look at the data.
> pdf(file = 'part9.scm.pdf', width = 8, height = 8)
> pairs(cbind(y, X[, -1]))
> dev.off()
null device
```



```
> 
> 
> print(sstotal <- sum(y * y))
[1] 453823.4
> print(syy <- sstotal.corrected <- sum((y - mean(y))^2))
[1] 68809.28
>  # Does the assignment and printing on one line, to be lazy.
>  # 'print' applies on 'syy'.
>  # Compare sst and syy!
>
```

```
> ##### BLOCK 2 #####
> # Let's fit a model with the intercept only.
> z \leftarrow lm.fit(x = X[, 'Const', drop = FALSE], y = y)
      # Selecting a single row or col will return a simple vector
      # by default; use 'drop = FALSE' will keep it a matrix.
      # 'drop' means dropping the dimension info.
> print(coef(z))
   Const
90.50851
> # We know this estimate should equal the mean.
> # Does it?
> print(mean(y))
[1] 90.50851
> # Calc SSR and SSE.
> print(ssr.Const <- sum(fitted(z) ^2))</pre>
                                          # SSR
[1] 385014.2
> print(sse.Const <- sum(residuals(z) ^2)) # SSE.
[1] 68809.28
      # This should equal syy. Does it?
> print(syy)
[1] 68809.28
> print(ssr.Const + sse.Const)
[1] 453823.4
      # This should equal sstotal. Does it?
> print(sstotal)
[1] 453823.4
> # Let's test the significance of this 'pure-intercept' model.
> print((ssr.Const / 1) / (sse.Const / (n - 1)))
[1] 257.3875
> print(qf(1 - alpha, 1, n - 1))
[1] 4.051749
      # Is the test statistic greater than the critical value?
      # Is it surprising to you that the intercept is significant?
> ##### BLOCK 3 #####
> # Let's add predictor 'Age'.
```

```
> z \leftarrow lm.fit(x = X[, c('Const', 'Age')], y = y)
> print(coef(z))
      Const
                    Age
128.6645573 -0.2753469
> # Calc SSR and SSE.
> print(ssr.ConstAge <- sum(fitted(z) ^2)) # SSR</pre>
[1] 385565
> print(sse.ConstAge <- sum(residuals(z) ^2)) # SSE.
[1] 68258.44
> print(ssr.ConstAge + sse.ConstAge)
[1] 453823.4
      # This should equal sstotal. Does it?
> print(sstotal)
[1] 453823.4
> # Let's test the significance of the coef for 'Age'.
> print(((ssr.ConstAge - ssr.Const)/ 1) / (sse.ConstAge / (n - 2)))
[1] 0.3631461
> print(qf(1 - alpha, 1, n - 2))
[1] 4.056612
      # Is the test statistic greater than the critical value?
> # Turns out to be insignificant.
> # Compare the sse's:
> print(sse.Const)
[1] 68809.28
> print(sse.ConstAge)
[1] 68258.44
      # The decrease is indeed small.
> # ssr.ConstAge - ssr.Const should be equal to
> # sse.Const - sse.ConstAge.
> # Is it?
> print(ssr.ConstAge - ssr.Const)
[1] 550.8396
> print(sse.Const - sse.ConstAge)
[1] 550.8396
      # Both are SSR(Age | Const)
>
> # Out of curiosity,
> # is the extra SS of 'Age' the same as the SSR of 'Age' alone?
> z \leftarrow lm.fit(x = X[, 'Age', drop = FALSE], y = y)
> print(ssr.Age <- sum(fitted(z) ^ 2))</pre>
```

```
[1] 379351.5
> # Compare the above with SSR(Age | Const).
> # The SSR of 'Age' alone is much larger than its extra contribution
> # on top of 'Const'.
> # Now, adding 'Age' on top of 'Const' is not significant.
> # Does 'Age' alone makes a significant model?
> sse.Age <- sum(residuals(z) ^ 2)
> print( (ssr.Age / 1) / (sse.Age / (n - 1)) )
[1] 234.3188
> print(qf(1 - alpha, 1, n - 1))
[1] 4.051749
      # Is the test statistic greater than the critical value?
>
> ##### BLOCK 4 #####
> # Does 'Const' and 'Age' as a group make a significant model?
> # The answer must be 'yes', given the preceding results.
> # But let's do a test anyway.
> # The things we need are already computed.
> print( (ssr.ConstAge / 2) / (sse.ConstAge / (n - 2)) )
[1] 127.0936
> print(qf(1 - alpha, 2, n - 2))
[1] 3.204317
>
> ##### BLOCK 5 #####
> # Keeping 'Const' and 'Age' in the model,
> # let's add 'Ed' and 'Ex0' at once.
> z \leftarrow lm.fit(x = X[, c('Const', 'Age', 'Ed', 'ExO')], y = y)
> print(coef(z))
       Const
                      Age
                                    Ed
                                                 Ex0
-221.0878571
                1.2300299
                             0.4737244
                                           1.0717904
> ssr.CAEE <- sum(fitted(z) ^ 2)
> sse.CAEE <- sum(residuals(z) ^ 2)
> print(ssr.CAEE + sse.CAEE) # This should be equal to 'sstotal'.
[1] 453823.4
> print(sstotal)
[1] 453823.4
```

```
> # Let's test the group.
> print( ((ssr.CAEE - ssr.ConstAge) / 2) / (sse.CAEE / (n - 4)) )
[1] 28.65531
> print(qf(1 - alpha, 2, n - 4))
[1] 3.214480
>
```